



MICROCOPY RESOLUTION TEST CHART
NATIONAL BUREAU OF STANDARDS-1963-A

Le

AIR FORCE

WEAPON SYSTEM COSTING: AN INVESTIGATION INTO CAUSE-EFFECT RELATIONSHIPS

By

Rosemarie J. Preidis

LOGISTICS AND HUMAN FACTORS DIVISION Wright-Patterson Air Force Base, Ohio 45433

> Mark Hollingsworth eral Dynamics Corporatio P.O. Box 748 Fort Worth, Texas 76101

> > June 1983

Approved for public release; distribution unlimited.

LABORATORY

AIR FORCE SYSTEMS COMMAND **BROOKS AIR FORCE BASE, TEXAS 78235**

83 06 07

昌

RESOURCES

NOTICE

When Government drawings, specifications, or other data are used for any purpose other than in connection with a definitely Government-related procurement, the United States Government incurs no responsibility or any obligation whatsoever. The fact that the Government may have formulated or in any way supplied the said drawings, specifications, or other data, is not to be regarded by implication, or otherwise in any manner construed, as licensing the holder, or any other person or corporation; or as conveying any rights or permission to manufacture, use, or sell any patented invention that may in any way be related thereto.

The Public Affairs Office has reviewed this paper, and it is releasable to the National Technical Information Service, where it will be available to the general public, including foreign nationals.

This paper has been reviewed and is approved for publication.

JOSEPH A. BIRT, Lt Col, USAF Technical Director Logistics and Human Factors Division

DONALD C. TETMEYER, Colonel, USAF Chief, Logistics and Human Factors Division

Unclassified

SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

REPORT DOCUMENTATION	ON PAGE	READ INSTRUCTIONS BEFORE COMPLETING FORM
REPORT NUMBER	2. GOVT ACCESSION NO.	t
AFHRL-TP-83-13		
L TITLE (and Subtile)	<u> </u>	5. TYPE OF REPORT & PERIOD COVERED
WEAPON SYSTEM COSTING: AN INVESTIGA	ATION INTO	Final
CAUSE-EFFECT RELATIONSHIPS		6. PERFORMING ORG. REPORT NUMBER
7. AUTHOR <i>IO</i>		8. CONTRACT OR GRANT NUMBER (s)
Rosemarie J. Preidis		F33615-79-C-0028
Mark Hollingsworth		
		LO DROOD AM ELEMENTE DROVECTE TASK
D. PERFORMING ORGANIZATION NAME AND ADDRESS General Pynamics Corporation		10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS
P.(). Box 148		62205F
Fort World, Texas 76101		17100021
		Lie DEPOND DATE
1. CONTROLLING OFFICE NAME AND ADDRESS		12. REPORT DATE
HQ Air Force Human Resources Laboratory (AFSC) Brooks Air Force Base, Texas 78235		June 1983
Drooks Air Force Dase, Texas (8255		13. NUMBER OF PAGES
		60
4. MONITORING AGENCY NAME & ADDRESS (if different	from Controlling Office)	15. SECURITY CLASS (of this report) Unclassified
Logistics and Human Factors Division Air Force Human Resources Laboratory		Unclassificu
Wright-Patterson Air Force Base, Ohio 45433		Pro Accused to the control of the co
wright-ratherson /til rotte base, Onto Wikks		15.a. DECLASSIFICATION/DOWNGRADING SCHEDULE
AND THE AND COLORS OF A PROPERTY OF A PROPER		L
6. DISTRIBUTION STATEMENT (of this Report)		
Approved for public release; distribution unlimited.		
•		
7. DISTRIBUTION STATEMENT (of this abstract entered in B	lock 20, if different from Rep	port)
	_	
8. SUPPLEMENTARY NOTES		
9. KEY WORDS (Continue on reverse side if necessary and ide	entife he black aug to 1	
	entry by Mack number) metric cost estimation	
•	on system acquisition pro	YPES
life cycle costing	, etc.ii acquieition pit	
·· / ·· · · · · · · · · · · · · · · · ·		
O. ANSTRACT (Continue on reverse side if necessary and ident	ife by black augher)	
Cause-Effect Cost Analysis was a technical	y) of mock number) effort that was terminated	after annovimately a year. Its oursees
was to establish an approach to costing aircraft av		
estimate end costs, but could provide design engi		•
requirements which, in turn, incur costs. A method		
consisted of two system design characteristics (tec		
selected system ownership requirements. The major		
was in defining and constructing the quantitative mea	•	• •
Ç		· · ·
Form 1473 EDITION OF 1 NOV 65 IS OBSOLETE		assified
l Jan 73		Man ab attention of the second
	SECTIONS OF AGGING	TION OF THIS DACE (The Date Francis)

SECURITY CLASSIFICATION OF THIS PAGE (When Date Entered)

Item 20 (Continued)	OF THIS PAGE (When Date En		
This paper describes the technical effort up to the point of termination. The avionics approach, as well as the cause-effect concept of weapon system costing, represents only an investigation into causal modeling and does not reflect validated techniques ready for use			
	1		
			<i>-</i>
			,
			•

Unclassified

SECURITY CLASSIFICATION OF THIS PAGE (When Date Entered)

FAR Z

WEAPON SYSTEM COSTING: AN INVESTIGATION INTO CAUSE-EFFECT RELATIONSHIPS

Ву

Rosemarie J. Preidis

LOGISTICS AND HUMAN FACTORS DIVISION Wright-Patterson Air Force Base, Ohio 45433

Mark Hollingsworth

General Dynamics Corporation P.O. Box 748 Fort Worth, Texas 76101

Reviewed by

William B. Askren
Chief, Acquisition Logistics Branch
Logistics and Human Factors Division

Submitted for publication by

Donald C. Tetmeyer, Colonel, USAF Chief, Logistics and Human Factors Division



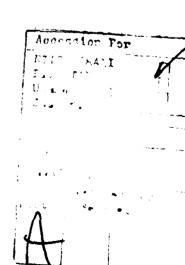


TABLE OF CONTENTS

		Page
ι.	INTRODUCTION	1
II.	BACKGROUND	1
III.	DISCUSSION	5
Footn	otes	21
Refer	ences	22
Appen	dix A: Development of Estimating Relationships	24
	LIST OF TABLES	
Table		Page
1.	Receivers in F-111 and F-16 Data Base	15
2.	Estimating Relationships	17
3.	F-15 Radar Receiver Test Case	18
4.	F-16 Radar Receiver Test Case	19
	LIST OF ILLUSTRATIONS	
Figur		Page
_		_
1.	Technology Described in Terms of Electronics Type	10
2.	Avionics Functional Density Increases with	10
3.	Complexity Scales Developed on Basis of F-111 and F-16 Transmitters and Receivers	13

1. INTRODUCTION

This paper summarizes the results of a study initiated by H. Anthony Baran, research psychologist, of the Air Force Human Resources Laboratory. The purpose of the study was to explore how a weapon system gencrates system ownership requirements in hopes of creating a new approach to weapon system costing. The new approach would involve cause-effect relationships between the phenomena that control the system design and the human and material requirements those phenomena percipitate. Unlike cost-estimating equations that estimate end costs based on technical experience, the causal relationships would provide insight into the circumstances which impact system ownership requirements that drive cost. Since the Department of Defense (DOD) is aware that the bulk of operation and support costs are incurred up front in system acquisition, it has emphasized the necessity for development of techniques for early and credible weapon system costing. Cause-effect relationships are among those techniques. The AFIRL study is introduced through an overview of how the DOD currently does its costing of weapon systems and the advantages and disadvantages to be associated with each method.

11. BACKGROUND

Parametric cost-estimating relationships are equations that attempt to describe mathematically the cost of an end item as a function of one or more variables. The equations are primarily used to develop probable cost estimates in the absence of hard data. The main premise behind cost estimating relationships (CERs) is that they are based on observations of what has happened in the past. The factors which prevailed

then should also hold during the estimated period. CERs are commonly derived in the following manner: candidate explanatory variables are normally selected as a result of interviews with engineers and manufacturers about the probable factors that impact end costs. Once the variables are identified to the best of the analyst's ability, their statistical properties are determined through multivariate regression analysis. If the equation is determined to be useful, i.e., if the statistical goals of the analyst are met, the equation is used.

Although it is not known where CERs were first developed, they have been used extensively by DOD since the early 1960s, as well as by private industry (Large, 1981, p. 2). Of the three basic methods used to estimate costs within DOD, parametric CERs are used most. There is good reason why CERs are popular in DOD. They provide what many Government analysts consider to be realistic cost estimates (Smith, 1971, p. 20). CERs also save manpower and time. For example, the accounting method, one of the three basic estimating methods, equates known units of costs to known units of output. It is an efficient method when there is sufficient cost information known about a design. In the early stages of weapon system acquisition, little is known about a piece of equipment other than minimum system characteristics and data such as engineering labor hours and labor ates. Other functional cost categories are almost nonexistent; this renders the accounting method inadequate. The cost analyst must resort to CERs to generate an estimate based on the available variables. An analyst may need to know what it will potentially cost to produce one fighter aircraft. The analyst who has no

cost proposal to evaluate will most likely use variables such as aircraft speed and airframe weight to predict the total number of direct
labor hours it will take to produce one aircraft. By multiplying the
estimated labor hours by the approximate labor rate, the analyst should
have a reasonable cost estimate based on minimal input.

The second method used in DOD to estimate costs is the engineering method. This involves building a cost profile of the aircraft from the ground up. Estimates from a cross-section of work segments (e.g., drafting, engineering, manufacturing) are consolidated into a project estimate. This can be a laborious process, depending on the end item that is being estimated. For example, estimating the cost of an airframe may involve 4,000 separate estimates (Poindexter, 19/6, p 22).

CERs can be used to generate a probable cost profile in short order since the majority of CERs are developed using computers. An analyst sitting at an interactive terminal can develop a functional relationship using a canned regression package in short order.

The biggest calling card for CBRs is that they are usoful in making cost estimates on fixed-wing aircraft, turbojet engines, missile systems, avionics components, and weaponry.

CERS are used on a variety of levels to estimate aircraft and weapon system costs. Typical CERs are those that derive cost estimates by type of aircraft structure, such as skin composition or machine plate; by functional cost elements, such as direct labor; and by acquisition phase. Airframe cost expressed as total direct labor hours as a function of airframe weight and speed has been used by DOD because of its estimating capability (Large, 1981, pp 3-8).

CBRs are used individually or in clusters depending on the estimation requirement. They are also found embedded in life cycle cost (LCC) models in which they perform an invaluable function; they generate values for cost elements, such as replacement spares or maintenance of technical orders, when no historical or comparable data are available.

LCC models used by the Air Force to estimate recurring or nonrecurring costs of aircraft systems rely heavily on CBRs. Examples include the LSC model, CACK model, DAPCA, and PRICE models².

Of late, CERs have been subject to some criticism. Estimating relationships that are based on physical and performance variables may now be inadequate to estimate costs of technologically advanced aircraft systems because of no historical experience. Experts have observed that "cost data collected on even the latest weapon systems represent not the cost of current technology but current cost of technology 5 to 10 years in the past. In 20 years, electronics have gone from tubes to microminiaturization. Materials technology is rapidly improving airframe construction. Motal structure components are being replaced by plastic components. Which, if any, historical technology is similar enough to any proposed system to allow valid design and credible cost?" (Haese, 1977, p. 34)

CERs are also not sensitive to design changes or to advanced technology. CERs estimate end costs. They do not consider the implications of a design's composition and its probable impact on the human and material requirements in the operating environment. DOD has intensified its efforts in procuring weapon systems that are better and

cheaper to operate and maintain³. To procure the best design at least cost without jeopardizing mission requirements requires cost analysis techniques whose components are sensitive to the drivers of system ownership requirements. Although there are many LCC models in use today that do satisfactory jobs (for what they were programmed to do), none contain well-defined causal relationships between system design characteristics and ownership requirements. Because of this, researchers are constantly striving for unique methods to analyze system designs in terms of their absolute effects on manpower, material, and costs. This suggests a technique that should have an important role in weapon system acquisition: causal modeling.

Causal modeling attempts to explain the potential determinants of effects. The technique requires that three conditions be met before a causal relationship may be inferred: covariation and time ordering must exist between variables, and the relationship must not be contaminated by unknowns whose effects may be significant.

III. DISCUSSION

In 1980-81, the Air Force Human Resources Laboratory conducted a study on cause-effect relationships between system design and system ownership requirements using causal modeling concepts. The purpose of the study was to develop a methodology to predict ownership requirements based on specific design inputs; the theory was that system design characteristics have a significant impact on the human and material requirements needed to operate and support the system in the field. The methodology, geared for aircraft avionics, would aid both systems

designers and support planners in assessing the impacts of alternative design configurations on ownership requirements. Two modifications to the methodology were also planned to extend its application to aircraft engines and missile systems, respectively. The study was divided into three phases to handle the aircraft avionics, engine, and missile system applications. Only phase one was completed. The other phases were terminated because of delays in technical progress. The results of phase one are presented in this paper.

A prototype methodology was established for avionics. The methodology consisted of a generic categorization of avionics equipment and a technique for analyzing these categories. The generic categories for avionics were transmitters, receivers, processors, sensors, displays, and controls. The technique consisted of a set of mathematical equations for selected ownership requirements as a function of two major design characteristics.

The generic categories seemed to encompass all of the representative electronic and nonelectronic functions that were to be associated with avionics equipment. Definitions for each of the categories are as follows:

Receivers - receive electromagnetic radiation including infrared.

This category includes receivers for radar, communications, and instrument landing systems. Electronically, the receiver functions include the circuitry for radio frequency amplifiers, detectors, mixers, local oscillators, and noise filters.

<u>Transmitters</u> - transmit electromagnetic radiation including laser transmissions. Electronically, the transmitter functions include power amplifiers, modulators, filters, mixers, and oscillators.

<u>Processors</u> - process data, signals or information. This category includes computers, signal converters, processors, and synchronizers. Specific amplifiers and power supplies packaged in separate line replaceable units (LRUs) are included in this group. The signal-processing circuitry of tactical communications and navigation, instrument landing systems, encoder/transponders such as identify friend or foe (IFF), and similar units are also within this category.

Sensors - gather and sense signals of electromagnetic radiation, motion, and pressure. This category includes such devices as antennas, gyros, accelerometers, air-data probes, pressure probes, vidicons, and other associated circuitry.

<u>Displays</u> - display information to the aircrew. This category includes the various readout and display devices ranging from mechanical digital readouts to cathode-ray tubes.

<u>Controls</u> - are the devices by which the aircrew puts information into the avionics system. These devices include knobs, switches, and keyboards. Generally, the control panels associated with different avionics systems will be in this category.

This categorization scheme appeared ideal for use on a new system early in the design stage at which point functions could be easily identified but LRUs could not.

The categories are comprehensive. Every avionics function falls within one of the categories. Consequently, an LRU or functional parts of an LRU can be categorized. The assumption that drives this categorization scheme is that the LRU cost and reliability values can be apportioned among the functions by the percentage of circuitry devoted to each function.

The next step was the development of a technique for analyzing these categories. A set of causal relationships (mentioned previously as mathematical equations) was developed for each generic category of avionics equipment. The relationships use complexity and technology indices as the primary inputs to predict specific ownership requirements. A technology index was used because it appeared to be a theoretically plausible predictor of avionics ownership requirements. It was assumed that the fundamental drivers of avionics cost and reliability would be governed by the size of the avionics box and its contents. The contents of the box are described by the kind of electronics which is, in turn, defined by function, technology age, and the amount of experience with that technology. A technology index and complexity index were developed for this purpose.

Technology age can be described in terms of electronic functions performed per unit weight. Such a relationship was developed by William A. Falkenstein of Ling-Temco-Vought (LTV) Aerospace Corporation in a paper given at the 33rd Annual Conference of the Society of Allied Weight Engineers, Inc. (SAWE) in 1974 (Falkenstein, 1974). This relationship was used as the technology index curve for this research and is illustrated in Figures 1 and 2.

The technology index indicates the density of electronic components. In the year 1945, one pound of avionics equipment contained approximately 10 vacuum tubes and their associated components. The term function denotes the activity (amplifying, gating, rectifying, etc.) of a single tube. Therefore, in 1945, avionics equipment performed approximately 10 functions per pound of electronics as indicated by point A of Figure 2. By 1955, improved technology led to the use of miniature tubes, and 1 pound of avionics contained approximately 20 miniature tubes. So, in 1955, 1 pound of avionics performed approximately 20 elementary functions (amplify, rectify, gating, etc.). This gives point B of Figure 2. As new technologies developed, the availability of medium scale integration (MS1), large scale integration (LS1), and very large scale integration (VLS1) allowed higher density electronics. Those give the points D, E, and F of Figure 2, respectively.

The technology index indicates the type of electronic components used rather than the sophistication of the system. In the previous paragraph, the word "function" referred to the simple operations of amplify, rectify, gate, etc., performed by a single vacuum tube. Elsewhere in this report, the word "function" will refer to a higher level operation, such as receive, transmit, and process. The major problem with the technology index is that no equivalent metric exists for translating vacuum tube functions into microelectronic functions and vice versa.

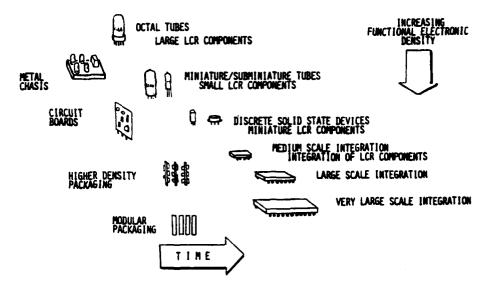


Figure 1. Technology Described in Terms of Electronics Type.

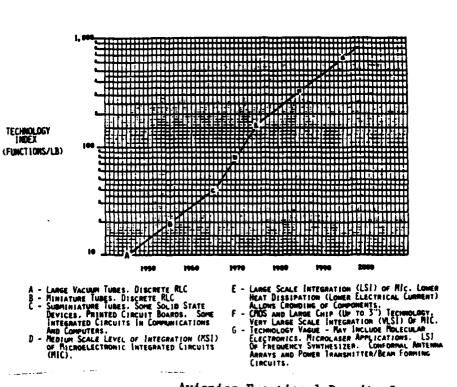


Figure 2. Avionics Functional Density Increases
With Advancing Technology

The technology index gives the vintage (period) of the electronics in the box. The complexity index is a measure of how much of this electronics is in the box. For a given vintage and kind of electronics, a measure of the amount of electronics (number of electronic functions) is weight. That is, if the period of the equipment determines the number of electronic functions per pound and the weight of the equipment is known, then the number of functions performed by the equipment is implied. The complexity index can be considered a scale of the number of electronic functions within the avionics subfunctions where: Complexity Index = Function Technology Index x Function 100 Circuitry

A complexity scale was developed for each avionics function category. Existing F-111 and F-16 avionics equipment was used to develop the complexity scale. For example, to develop the complexity scale for receivers, all LRUs with a receiver function were grouped by subsystem type and were listed along with the LRU weight, the percentage of the circuitry within the LRU devoted to the receiver function, and the technology index of each LRU. Engineering judgment was used to establish the technology indices for the LRUs. All the LRU weights were normalized for technology. The LRU weight percentage devoted to the receiver function became a point on the complexity scale.

The complexity index developed for the receiver category is shown in Figure 3. Also shown are preliminary complexity scale values for transmitters. To use the scale for a piece of avionics equipment that has a

receiver function, one compares the performance characteristics of the new equipment with the performance of receivers within the subsystem category of points existing on the receiver complexity scale.

Other variables were considered for constructing the causal relationships, the most critical being the amount of built-in test (BIT) circuitry in an LRU because of its support concept implications. The amount of BIT circuitry also impacts the complexity index of an avionics component. All the points on the complexity scale were adjusted to eliminate BIT potentially. It was assumed that if an avionics function contained BIT circuitry that could detect and isolate 95% of the faults, its complexity index would be increased by 10%. The gross 10% estimate was derived from an analysis of BIT circuitry within F-16 avionics equipment. The complexity index would be reduced by 10% if a piece of avionics equipment had 95% fault isolation capability. Again, subjective judgment figured largely in defining the BIT percentages and the complexity scale.

Next, causal relationships for the receiver function were developed. The relationships were developed to analyze the functional rather than the physical unit such as an LRU. The reasoning was that, during the early design phase, the physical packaging of equipment may not be known. Analysis of the functional unit allows this technique to be used early in the acquisition phases. These relationships were derived through multivariate regression.

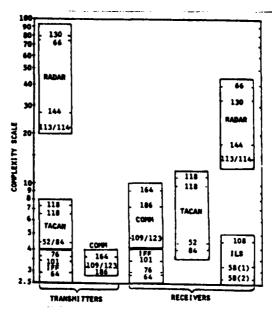


Figure 3. Complexity Scales Developed on Basis of F-111 and F-16 Transmitters and Receivers.

The relationships were developed from an F-111 and F-16 equipment data base containing 20 data points for the generic receiver category shown in Table 1. Data were collected at the physical unit level and were adjusted to reflect the functional unit by multiplying each data point by the percent of the LRU dedicated to the function. BIT circuitry was also subtracted from the functions using engineering judgment. The relationships that were developed are shown in Table 2. The appendix contains explanations of the predicted variables. An analysis of variance table for each equation is also included in the appendix.

A set of the relationships was tested on an F-15 radar receiver, a unit that was not in the data base. The predicted values came close to the actual F-15 values. The set of receiver relationships and test results are shown in Table 3. The relationships were also tested on an F-16 radar receiver. Those results are shown in Table 4.

The concept governing these relationships appears valid. The relationships were developed for individual categories of electronic and nonelectronic functions so that predictions could be based on trends of similar functions. Second, trend comparisions were made at similar technology levels through the technology and complexity indices. Third, the physical packaging of equipment did not appear to influence the basic predictions. Last, the primary inputs of the causal relationships, the complexity and technology indices, appeared to provide some design sensitivity to selected ownership requirements.

TABLE 1

Receivers in F-111 and F-16 Data Base

ATTACK RADAR	TERRAIN FOLLOWING RADAR	RADAR ALTIMETER
AN/APQ-113 APQ-114 APQ-130 APQ-144 APG- 66	AN/APQ-128	AN/APN-167
<u>IFF</u>	ILS	TACAN
AN/APX~ 64 APX~ 76 APX~101	AN/ARN- 58 ARN-108	AN/ARN- 52 ARN- 84 ARN-118 (F-111 ARN-118 (F-16)
HF	VHF	UHF
AN/ARC-123	AN/ARC-186	AN/ARC-109 ARC-164

The methodology could be modified for application to engines and missile systems. The basic analysis of aircraft engines would be performed at the function level. The generic engine functions would include compressor, fuel monitoring, accessory gearbox, turbine and augmentor. Complexity and technology indices would be developed in a similar manner. The complexity index could indicate the number of piece parts in a function. A technology index for engines developed by Rand Corporation several years ago would be incorporated into the methodology. In addition, a technique for considering the metallurgical composition of the engine would be developed (assuming that metallurgical composition of engines is a factor in driving certain ownership requirements).

The avionics methodology could almost be directly applied to missile systems. It was determined that the same aircraft avionics categories could work for missiles since aircraft avionics and missile avionics have similar functions even though the missions differ. Only four generic functions would be needed to describe missile avionics: sense, transmit, receive, and process. The complexity and technology indices would be the same (i.e., the missile indices would be constructed in the same manner as the aircraft avionics indices).

Although this study was exploratory, it did focus attention on several system design characteristics that indicated reasonable causal relationships with selected system ownership requirements. Research in this area could prove invaluable to DOD. Causal relationships could enable system designers and support planners to anticipate the impacts

TABLE 2

ESTIMATING RELATIONSHIPS

LRUs Per Function

LPF = .2062 + .1044(complexity)

SRUs Per Function

SPF = 121.4(complexity)·2984(technology)-.7740

LRU Unit Cost

UC = (10019.6)(.98354)technology(1.10701)complexity

SRU Unit Cost

 $SUC = (253.43)(complexity) \cdot 7267$

Mean Time Between Defective Removals

MTBDR = (2144)(1.01962)technology(.91818)complexity(.22292)utility

Bench Check Serviceable Elapsed Time

BSET = .2267(complexity).6428

Bench Check on Repair Elapsed Time

BCRT* = (FIAT + FIXT)/(efficiency)

where FIAT = fault isolate and test time = .3237(complexity).6650

*(see p. 29 for description of other components of this equation).

Integrated Test Adaptor Cost

ITAC = 3919(complexity).4401

Test Software Cost

SWC \approx 9867 + 1431(complexity)

Technical Order Pages

TOP = 69.64 (complexity).4366

TABLE 3

F-15 Radar Receiver Test Case

Inputs	F-15		
Complexity Technology Utility	43 120 1		
Outputs	F-15 Estimate	F-15 Actual	
Mean Time Between Defective Removal	125	130	
Unit Cost	108,246	99,000	
Bench Check Servicable Elapse Time	3.8	?	
Bench Check & Repair Elapse Time	5.92	7.4	

TABLE 4
F-16 Radar Receiver Test Case

		F-16 RADAR RECEIVER SUBFUNCTION	
PREDICTED SUPPORT VARIABLE	IMPUTS	PREDICTED	ACTUAL
PARTITIONING			
- LRUS/SUBFUNCTION	COMPLEX	0.79	0.63 6.93
- SRUS/SUBFUNCTION	COMPLEX/TECH	0.25	9.73
COST	anner an famous	68167	60121
- LRU UNIT COST	COMPLEX/TECH	3759	6607
- SRU UNIT COST	COMPLEX	3/37	
ON-EQUIPMENT MAINTENANCE	COMPLEX/TECHNO/UTILITY	200	219
~ MIBDR ~ REMOVE AND REPLACE TIME	I FIC/BIT/CONSTANTS	2.06	1.80
- CANNOT DUPLICATE TIME	BIT/CONSTANTS	0.92	1.30
- ON-EQUIPMENT REPAIR	I FIC/BIT/CONSTANTS	1.39	
- CREW SIZE	BIT	2.2	2.2
OFF-EQUIPMENT MAINTENANCE			
- BENCH CHECK SERVICEABLE RATE	I FIC/# LRUS/I FI ERRORS	.136	.122
- BENCH CHECK SERVICEABLE	COMPLEX	3.68	3.30
ELAPSED TIME]
- BENCH CHECK & REPAIR	COMPLEX/CONSTRUCT/TECH	5.73	5.60
ELAPSE TIME	1	***	1
- REPAIR MATERIAL COST	COMSTANT	¥7	
SUPPORT EQUIPMENT		20069	19987
- ITA UNIT COST	COMPLEX	68111	77358
- SOFTMARE COST - TEST STATION MAN/ON	COMPLEX		
- TEST STATION ATBMA	CONSTANT	0.40	0.40
- TEST STATION MIDMA	CONSTANT	14.90	14.90
FACTOR	COMBIANT	•	•
- TEST STATION MATERIAL COST/OH	CONSTANT		1 _
TECHNICAL ORDER PAGES	COMPLEX	352	371
ISMINITURE GUREN LAGES	CONTLEA		3/4

of advanced technological designs on human and material resources through an understanding of the interrelationships of design, technology, and system ownership requirements.

FOOTNOTES

 $^{\rm I}$ Statistical goals usually include the following: establishing an ${\rm R}^2$ value; setting a standard error of the estimate percentage of mean response; an alpha value for statistical significance; and reviewing residuals for discernible patterns. These criteria should help the researcher decide whether a useful equation has been developed.

²LSC stands for Logistics Support Cost Model; CACE is Cost Analysis Cost Estimating; DAPCA is Development and Production Costs of Aircraft; PRICE is Programmed Review of Information for Costing and Evaluation.

³DOD has published numerous documents that stress the procurement of weapon systems that will prove to be economical to operate and support in the field. Such documents include DOD Directive 5000.2 Major System Acquisition Process; DOD Directive 5000.39, Acquisition and Management of Integrated Logistics Support for Systems and Equipment, and DOD Directive 5000.4, OSD Cost Analysis Improvement Group.

⁴AFHRL had assistance from General Dynamics Corporation, Fort Worth and Convair Divisions under contract number F33615-79-C-0028.

⁵Phase two, the engine study, was aborted to avoid duplication of effort in light of Rand Corporation's headway in engine cost estimating techniques. Phase three, the missile application, was ended because the contractor could not develop costing relationships for missiles. Despite unresolved technical difficulties, phase one was permitted to run its natural course since its constructs and the preliminary relationships for avionics costing had the strong appeal of "gut level" credibility. The difficulties, which undermined that credibility, lay in: 1) defining and quantifying the measures of avionics technology and complexity; and 2) proceduralizing the definition of avionics functions over time.

⁶John M. Jermier writes that one need not rule out all possible causes in a causal inquiry if there is "sufficient reasonableness" to expect a relationship between two or more variables; see, e.g., his article "Causal Analysis in the Organizational Sciences and Alternative Model Specifications and Evaluation," The Academy of Management Review, III (April 1978), pp. 326-37.

⁷The time-of-arrival equation was developed specifically for military jet engines. It is essentially a technology index composed of engine characteristics such as turbine inlet temperature, total pressure, weight, thrust and fuel consumption. J.L. Birkler, et al, describe this equation in <u>Development and Production Cost Estimating Relationships for Aircraft Turbine Engines</u> (Sarta Monica, California: Rand Corporation, 1982).

REFERENCES

- Asher, H. B. <u>Causal Modeling</u>. Quantitative Applications in the Social Sciences. Beverly Hills, California: Sage Publications, Inc., 1976.
- Birkler, J. L., Garfinkle, J. B., & Marks, K. B. <u>Development and Production Cost Estimating Relationships for Aircraft Turbine Engines</u>.

 Rand Note. Santa Monica, California: The Rand Corporation, 1982.
- Blalock, H. M. Jr., ed. <u>Causal Models in the Social Sciences</u>. Chicago: Aldine Publishing Company, 1971.
- Blanchard, B. S., & Fabrycky, W. J. <u>Systems Engineering and Analysis</u>. Englewood Cliffs, New Jersey: Prentice-Hall, Inc. 1981.
- Cheslow, R. T., & Dever, J. R. "Acquisition Costing in the Federal Government," <u>Defense Systems Management Review</u>, II (Autumn, 1979), pp. 7-16.

- Davis, W. R., & Wysowski, J. R.. "A Summary and Analysis of the Logistics Support Cost Model Application to the ACF/F-16 Weapon System Acquisition." Unpublished master's thesis. LSSR 6-79A, AFIT/SL, Wright-Patterson AFB, Ohio, September 1978.
- Falkenstein, W. A. "Methodologies for Predicting Avionic System Capability and Weight in CTOL and VTOL Fighter/Attack Aircraft 1975 to 1995." Paper presented at the 33rd Annual Conference of the Society of Allied Weight Engineers, Inc., Fort Worth, Texas, May 6-9, 1974.
- Haese, E. J. "How Useful are Historical Costs?" Air Porce Comptroller, XI (July 1977), pp. 34-36.
- Hitch, C. J. "Cost Estimating in the Department of Defense," <u>Armed Porces Comptroller</u>, X (March 1965), pp. 3-6.
- Hollingsworth, M. "Cause Effect Cost Analysis. Unpublished technical report, General Dynamics Corporation, Fort Worth Division, 1981.
- Large, J. P. <u>Development of Parametric Cost Models for Weapon Systems</u>. The Rand Paper Series. Santa Monica, California: The Rand Corporation, 1981.

- Marks, K. E., Massey, H. G., and Bradley, B. D. An Appraisal of Models

 <u>Used in Life Cycle Cost Estimation for USAF Aircraft Systems</u>. R
 2287-AF. A Project Air Force report prepared for the United States
 Air Force. Santa Monica, California: The Rand Corporation, 1978.
- Marks, K. E., Massey, H. G., Bradley, B. D., & Lu, J. "A New Approach to Modeling the Cost of Ownership for Aircraft Systems." R-2601-AF. A Project Air Force report prepared for the United States Air Force. Santa Monica, California: The Rand Corporation, 1981.
- Neuman, Y. "Assessing Causality in Organizational Studies: The Path-Analytic Approach." The Academy of Management Review, III (April, 1978), pp. 366-69.
- Ostwald, P. F. Cost Estimating for Engineering and Management.

 Englewood Cliffs, New Jersey: Prentice-Hall, Inc., 1974.
- Poindexter, E. O. "Cost for Decision Making: A Synopsis." <u>Armed Forces Comptroller</u>, XXI. (January, 1976), pp. 16-23.
- Rice, D. B. "Cost Analysis in DOD." <u>Armed Forces Comptroller</u>, XIII. (January 1968), pp. 18-21.
- Smith, A. T., "'Should Cost'" <u>Armed Forces Comptroller</u> XVI. (Winter, 1971), pp. 20-3.

APPENDIX A DEVELOPMENT OF ESTIMATING RELATIONSHIPS

Partitioning Variables

The partitioning variables help to indicate how the design engineers will partition the electronics. Although this can vary from one design engineer to another, it was reasoned that the complexity and technology of the equipment can explain some of the variation in packaging.

LRUs per function (LPF) is derived from regression analysis:

LPF = .2062 + .0144(complexity)

SRUs per function (SPF) was derived from regression analysis:

SPF = 121.4(complexity).2984 (technology).7740

Cost Variables

The significance of the cost variables is clear. Historical unit cost data will show a great deal of variation for some LRUs. For instance, a spare part produced during a production run can cost as little as one-fifth the cost of the same spare produced as a separate run. This is due to the large initial cost of a run. Furthermore, some "hot mockups" are purchased for training purposes and are much more expensive than the normal unit.

This cost variation will occur for any avionics system and is not adequately reflected in the Air Force Logistics Command data (e.g., KO51 records). In order to get valid average unit cost data, it was necessary to average several purchase records for each LRU.

The LRU unit cost (UC) was derived from regression analysis:

UC = avg. LRU unit spares cost (FY75\$)

= (10019.6)(.98354) technology (1.10701) complexity

The SRU unit cost (SUC) was derived from regression analysis:

SUC = avg. spares unit cost of the receive/transmit

function (FY75\$)

= (253.43)(complexity).7267

On-Equipment Maintenance Variables

The on-equipment maintenance variables depend greatly on the builtin test and fault isolation capabilities of avionics equipment.

Mean time between defective removal (MTBDR) was derived from regression analysis:

MTBDR = avg. # of flight hours between removals of a defective unit

where

utility = 0 for communications equipment: and
1 otherwise. The variable, utility, accounts for
variations in MTBDR for certain equipment which may have greater usage
demands placed on them, which, in turn, precipitates higher than
expected removals.

The elasped time for remove and replace, cannot duplicate, and onequipment repair are derived from estimates of the times required to set up, verify, troubleshoot, remove and replace, repair, LRUs and close up the unit.

The set-up time is the time required (1) to access the LRU, (2) to connect power and cooling air or start the auxilliary power unit (APU), and (3) to position work stands. Therefore,

set-up time = access time + electric power time + cooling air time +
maintenance stand time

where

These time estimates were obtained from analysis of F-16 and F-111 maintenance actions.

Verify time is the elapsed time required to functional check the system after LRU replacement.

Verify time = [.44 hr if equipment has BIT capability; and .8 hr otherwise

The verify times are obtained from analysis of F-16 and F-111 maintenance actions.

Troubleshoot time = .33(FIC) + .8(1-FIC) where

FIC = decimal fraction of the time that built-in test/self-test can fault isolate to the LRU level.

This equation is obtained by comparing F-111 times (very little built-in test/self-test circuitry) with F-16 times (.95% fault isolation capabilities).

Remove and replace LRU time is the elapsed time required to remove and replace the faulty component.

Remove and replace LRU time = set-up time + troubleshoot time + remove and replace LRU time + verify time + close-up time

On-aircraft repair time = .5(set-up time) + .5(troubleshoot time) + repair time + .5(verify time) + .5(close-up time)

Cannot duplicate time = .5(set-up time) + verify time + .5(close-up time)

Crew size = 2.2 if the equipment has BIT capability; and 2.4 otherwise.

The two values for crew size are averages from F-16 and F-111 data, respectively.

Off-Equipment Maintenance Variables

The bench check serviceable rate (BSCR) can be predicted theoretically.

BSCR = decimal fraction of time a good LRU is removed

= (1-FIC) (1-1/NLRU) + FIE,

where

FIC = decimal fraction of the time which the equipment identifies
an LRU to be had.

and

FIE = decimal fraction of time which the equipment improperly identifies an LRU to be bad when in fact the LRU is servicable

NLRU = the number of LRUs in the function.

Both FIC and FIE should be in specification of future avionics equipment. The value of NLRU can be estimated with the help of the partioning variables.

Bench check serviceable elapsed time (BSET) was derived from regression analysis:

BSET = .2267(complexity).6428

Bench check and repair elapsed time (BCRT) is:

BCRT = (FIAT + FIXT)/(efficiency)

where

FIAT = fault isolate and test time = .2267(complexity).6418 was derived from regression analysis:

FIXT = fix time

.25 for remove & replace of plug-in model
.33 for discrete components (2-4 pins)

-47 for MSI components (14 pins)
.60 for LSI components (24 pins)
.73 for VLSI components (36 pins)

and

efficiency = .6667 is an adjustment factor which accounts for the
skill differences between commercial maintenance
personnel and Air Force intermediate shop maintenance
personnel. (Contractor's subjective estimate).

Repair material cost (RMC) was derived from historical averages;

RMC = \$47

The average replacement part cost was \$43, and 10% was added for miscellaneous material cost.

Support Equipment Variables

Integrated Test Adaptor Cost was derived from regression analysis:

ITAC = 3919(complexity).4401

Test Software Cost (SWC) was derived from regression analysis:

SWC = 4867 + 1431(complexity)

The Test Station Cost (TSC) was averaged from F-16 test stations:

TSC = 800,000

The Test Station Man Hours (TSMH) is the average number of manhours per test station repair:

TSMH = .4

The Test Hean Time Between Maintenance Action (TMTBMA) is the average test time between test station repairs.

TMTBMA = 14.9

The Technical Order Pages (TOP) was derived from regression analysis:

TOP = 69.64(complexity).4366

LRUS PER FUNCTION

LPF= .2062 + .0144(complexity)

123456789	OBSERVATIONS 8 9 11 15 16 17 19 20 21	CROSS F	EF: DATA 718A0 718A0 716A0 73CA0 73KB0 73KB0 73BD0 73PK0 73VA0 74AB0	PT	DISK RC
			THE INPUT	VARIABLE	S ARE
И	LRUS∠SUBF Y	COMPLE X 1	EX N		
1	0.20	4.40	1		
2	0.20	4.00	2		
3	0.30	11.00	÷		
4	0.30	2.00	4		
5	0.50	8.40	5		
6	0.40	16.20	ē.		
7	1.00	37.20	ï		
8	0.40	20.00	8		
9	0.70	45.40	9		

(continued p. 33)

LRUs PER FUNCTION(Continued):

	OF INPUTS FOR O ENCE LEVEL = 0.		1+b* X 1+ c*X2	
COPPELATION MATE: 1.00000 0. 0.85070 1.	.85070	olamn/		
CONFIDENCE OF VAI 18.33365015	RIABLE 1 EN	XT ITERATION : TERED IS	0.995982302	F =
VAPIABLE 1 (CONSTANT	COEF 0.014428813 3 0.206208705	STAND ERR .36981 E-0 3	NAME COMPLEX	
ANOVA TABLE SOURCE	DF	SS	· · -	OVERALL F
TOTAL PEGPESSION PESIDUAL	8 Ø 1 Ø 7 Ø	.542222222 .392399 5 94 .14982 26 29	0.392399594 0.021403233	18.303
THE OVERALL COHE THE PER CENT OF	IDENCE IS 0 ERROR EXPLAINED	.9959823 02 IS	72. 368777 5	
tO OTHER VARIABL THE HVERAGE 1 ER THE AVERAGE ERRO	ROR IS 25.2 R IS 0	:3 .109902749		
085ERV EST 0.27 2 0.26 3 0.36 4 0.24 5 0.33 6 0.44 7 0.74 8 0.49 9 0.86			% ERROR -34.85 -31.96 -21.64 21.64 34.52 -9.99 25.70 -23.70	NAME 718A0 71EA0 71AA0 73CA0 73KB0 73BD0 73PK0 73YA0 74AB0
	PARTIAL F STAT	CONF.	NAME	
1	18.33366014 0	995982302	COMPLEX	

SRUS PER FUNCTION

-.7740

.2984 -.
SPF = 121.4(complexity) (technology)

	OBSERVATIONS	CROSS	REF:	DATA	PΤ	DISK
1	1		74	ABOR	•	2-41.
2	2			AAOR		
3	3			AAOT		
4	4			BAOR		
5	5		63	AAOR		
6	6 .		63	AROT .		
7	7		62	CAOR		
ខ	8		62	CAOT		
9	9		74	ACOT		

THE INPUT VARIABLES ARE

	SRU/SUBF	COMPLEX	TECH	40
N	Υ	X 1	Х 2	Н
1	7.70	46.00	135.00	1
2	4.00	3.40	135.00	2
3	4.00	3.40	135.00	3
4	6.40	4.50	80.00	4
5	9.10	9.00	135.00	5
6	3 .90	3,60	135.00	6
7	5.60	7.20	135.00	7
8	2.40	3,10	135.00	8
9	9.00	83.70	135.00	9

(continued p. 35)

SRUs PER FUNCTION(Continued):

	ION OF INPUTS FO FIDENCE LEVEL =	OR OPTION 2 Y= 0.5	•a*X1†b*X2†c	
1.00000 0.17930	ATRIX (Y is las 0.17930 0.7 1.00000 -0.1 -0.15429 1.0	'5351 15429		
CONFIDENCE OF 9.195523859	VARIABLE 1		0.981384752	F =
VARIABLE 1 CONSTANT	COEF 0.278613247 2.970758096	STAND ERF 0.091878421	R NAME COMPLEX	
ANOVA TABLE SOURCE	Di ^r	SS	MS	OVERALL F
TOTAL REGRESSION RESIDUAL	8 1 7	1.630751505 0.925911042 0.704840463	0.925911042 0.100691495	
THE OVERALL C THE PER CENT	ONFIDENCE IS OF ERROR EXPLAIN	0.9 813 8475 2 NED IS	56. 77818105	
CONFIDENCE OF 1.501906695		- NEXT ITERATION ENTERED IS	0.734026137	F =
VARIABLE 1 2 CONSTANT	COEF 0 298436377 -0.774007557 121.4069733	STAND ER 0.090213657 0.631573246	R NAME COMPLEX TECHNO	
ANOVA TABLE SOURCE	ŋ. .	, \$\$	ms	OVERALL F
TOTAL REGRESSION RESIDUAL	8 2 6	1.630751505 1.067022449 0.563729056	0.533511224 0.093954843	
THE OVERALL O	ONFI)ENCE IS OF ERROR EXPLAI	0.946703889 HED IS	65.43133309	

SRUs PER FUNCTION(Continued):

THE AVE	R VARIABLES CAN RAGE % ERROR IS RAGE ERROR IS	16.03	917246050		
088ERV 12 33 4 5 6 7 8	8.54 3.93 3.93 6.40 5.25 3.99	4.00 4.00 6.40 9.10 3.90 5.60	ERROR -0.84 0.07 0.07 -0.00 3.85 -0.09 0.69 -1.42	% ERROP -10.94 1.85 1.85 -0.00 42.31 -2.40 12.30 -59.14 -13.48	NAME 74ABOF 65AAOF 65AAOF 63AAOF 63AAOF 62CAOF 62CAOF 74ACOT
PARTIAL VA	SIGNIFICANCE OF PARTIE	OF VARIABLES AL F STAT	IN REGRESS	NOIS MAME	
2		356800 0. 906695 0.			
PARTIAL	SIGNIFICANCE (OF VARIABLES	NOT IN RE	GRESSION	
VAR	PARTIAL F	STAT	CONF	NAME	

•••	********	techno	1000	comple	witu
uc :	= (10019.6)((1.1070	•	XILY
1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16	OBSERVA	TIONS CROSS 2 3 4 7 9 11 12 13 14 15 16 17 18 19 20 21	REF: DAT 63AA0 63AA0 62CA0 65AA0 71EA0 71CA0 71CB0 71SA0 73CA0 73CA0 73CA0 73CA0 73CA0 73CA0 73CA0 73CA0	A PT	DISE. (
			THE INPUT	VARIAB	LES ARE
н	COST Y	TECHN X 1	IO COMPI X 2	EX N	
1	6671.00	60.00	4.70	1	
2	2461.00	135.00	9.00	2	
3	2538.00	135.00	7.20	3	
4	2818.00	135.00	3.40	4	
5	2157.00	80.00	4.00	5	
6	2454.00	135.00	11.00	6	
7	7660.00	50.00	2.70	7	
8	7680 .0 0	50.00	3.20	3	
9	3637.00	80.00	4.50	9	
10	2559.00	60.00	2.00	10	
11	10168.00	60.00	8.40	11	
12	25528.00	40.00	16.20	12	
13	22000.00	40.00	16.20	13	
14	167357.00	60.00	37.20	14	
15	54084.00	50.00	20.00	15	
16	60121.00	135.00	40.90	16	

(continued p.38)

LRU UNIT COST(Continued):

	N OF INPUTS FO DENCE LEVEL ≃	OR OPTION 3 YO 0.5	=a*b†X1*c†X2	
CORRELATION MAT 1.00000 0.07440 -0.40427	RIX (Y is los 0.07440 -0.4 1.00000 0.3 0.84868 1.0	10427		
CONFIDENCE OF V 36.04592652	ARIABLE 2	NEXT ITERATION ENTERED IS	0.999891997	F =
VARIABLE 2 CONSTANT	COEF 1.102566394 2715.630863	STAND ER 0.016263054	R NAME COMPLEX	
ANOVA TABLE SOURCE	DF	SS	MS	OVERALL F
TOTAL REGRESSION RESIDUAL	15 1 14	28,15187104 20,27658084 7,875290200	2 0. 27658084 0. 562520729	
THE OVERALL CON THE PER CENT OF	ERROR EXPLAIN	ED 18	72.02569524	
CONFIDENCE OF V 47.5638315	ARIABLE 1	NEXT ITERATION ENTERED IS	0.999943971	F =
VARIABLE 1 2 CONSTANT	COEF 0.983542908 1.107011504 10019.6045	STAND ER 2.40610E-03 7.84088E-03	R NAME TECHNO COMPLEX	
AMOVA TABLE SOURCE	DF	SS	MS	OVERALL F
TOTAL REGRESSION PESIDUAL	15 2 13	28.15187104 26.46144344 1.690427604	13.23072172	
THE OVERALL CON THE PER CENT OF			93.99532769	

LRU UNIT COST(Continued):

THE AVE	ER VARIABLE ERAGE % ERR ERAGE ERROR		D .90 3273.314255		
08SERV 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15	EST Y 596.867 2662.67 2217.39 1506.83 3989.55 3263.03 5750.78 6050.66 4197.59 4536.84 8692.28 26782.28 26782.28 162520.68 33385.57 68197.73	TRUE Y 6671.00 2461.00 2538.00 2157.00 2454.00 7660.00 3637.00 2559.00 10168.00 25528.00 25528.00 25528.00 267357.00 54084.00 60121.00	ERROR 701.14 -201.67 320.61 1311.17 -1832.55 -809.03 1909.22 1629.34 -560.59 -1977.84 1471.85 -1254.28 -471.85 -1254.28 -4836.32 20698.43 -8076.73	% ERROR 10.51 -8.19 12.63 46.96 -32.97 24.92 21,22 -15.41 -77.29 14.48 -4.91 -21.89 38.27 -13.43	NAME 63AA0 63AA0 62CA0 62CA0 71EA0 71CA0 71CB0 71CB0 73CA0 73KB0 73KB0 73KB0 73KB0 73KB0
		NCE OF VARIAB ARTIAL F STAT			

0.999943971 0.999997851 TECHNO COMPLEX

47.5638315 168.1143938

SRU UNIT COST

.7267

SUC = (253.43)(complexity)

	OBSERVATIONS	CROSS	REF:	DATA	ÞΤ	I
1	1		74	ABØR		
2	2		651	AAQR		
3	3		65	AAOT		
4	4		71	BAOP		
5	5		631	AAAR		
6	6		63	AAGT		
7	9		74	ACOT		

THE INPUT VARIABLES APE

	ASRU COST	COMPLE	TECH	ł0
Н	Y	X 1	X 2	N
1	6607.00	46.00	135.00	1
2	726.00	3.40	135.00	2
3	726.00	3.40	135.00	3
4	573.00	4.50	80.00	4
5	769.00	9.00	135.00	5
ε	769.00	3.60	135.00	6
7	5075.00	83.70	135.00	7

(continued p. 41)

SRU UNIT COST(Continued):

---- REGRESSION OF INPUTS FOR OPTION 2 Y=a*X11b*X21c ------CONFIDENCE LEVEL = 0.5 COPRELATION MATRI ((Y is last column) 1.00000 0.34174 0.94566 1.00000 0.34696 0.24174 0.94566 0.34696 1.00000 ----- NEXT ITERATION ---------CONFIDENCE OF VAR!ABLE 1 ENTERED IS 0.998036404 F = 42.28914797 VARIABLE COEF STAND ERR NAME 0 726683968 0.111745775 COMPLEX CONSTANT 253.429932 AHOVA TABLE \mathbf{D}^{17} SOURCE MS OVERALL F TOTAL 6 6.39897142 PEGRESSION 5.722392152 5.722392152 42,289 1 PESIDUAL 5 0.1353158**5**4 0.676579268 THE OVERALL CONFIDENCE IS: 0.998036404 THE PER CENT OF ERROR EXPLAINED IS 89,42674966 NO OTHER MARIABLES CAN BE ADDED THE AVERAGE : EPP OF IS THE AVERAGE EPROF IS 29.12 681.8478735 OBSERV EST Y TRUE Y ERROR % ERROR NAME 38.03 4094.09 6607.00 2512.91 74AERON 1 726.00 726.00 573.00 769.00 15.05 616.70 65AAOP 109.30 616.70 756.03 109.30 15.05 65AAOT -183.03 -31.94 71BA0P 1251.10 -62.69 -482.10 63AAOF 638801 642.86 769.00 126.14 16.40 6325.16 5075.00 -1250.16-24.63 74AC01 PARTIAL SIGNIFICANCE OF VARIABLES IN REGRESSION VAF PARTIAL F STAT NAME CONF. 1 43.28914796 0.995825840 COMPLEX PARTIAL SIGNIFICANCE OF VARIABLES NOT IN REGRESSION PARTIAL F STAT CONF MAR NAME 0 655023851 0.459213480 TECHNO

MEAN TIME BETWEEN DEFECTIVE REMOVALS

technology complexity utility

MTBDR = (2144)(1.01962) (.91818) (.22292)

	OBSERVATIONS	CROSS	REF:	DATA	РΤ
1	_		61A		
2	1 2 3		63AI		
3	3		63AI		
4	4		6201	10	
5	5		65A		
6	6		6586		
23456789	4 5 6 7 9		65Af		
8			71E	10	
	10		7126		
10	11		7186	90	
11	12		7106	90 F	
12	13		710		
13	14		71B		
14	15		73Cf	10	
15	16		73K8	30	
16	17		7381	00	
17	18		73.J0	0	
18	19		73F)	(Ø	
19	20		73Vf	10	
20	21		7488	30	

THE INPUT VARIABLES ARE

N	MTBDR Y	TECHNO × 1	COMPLEX X 2	X 3	LITY
1	307.00	60.00	4.70	2.00	1
2	150.00	60.00	4.70	2.00	2
3	456.00	135.00	9.00	2.00	3
4	654.00	135.00	7.20	2.00	4
5	905.00	50.00	2.90	1.00	5
6	1010.00	50.00	2.40	1.00	6
7	2361.00	135.00	3.40	1.00	7
8	1467.00	80.00	4.00	1.00	8
9	4581.00	135.00	9.70	1.00	9
10	4177.00	135.00	11.00	1, 66	1.0

(continued p. 43)

MEAN TIME BETWEEN DEFECTIVE REMOVALS(Continued):

11	1608.00	50.00	2.70	1.00	11
12	1173.00	50.00	3.20	1.00	12
13	1964.00	80.00	4.50	1.00	13
14	490.00	60.00	2.00	2.00	14
15	202.00	60.00	8.40	2.00	15
16	183.00	40.00	16.20	1.00	16
17	128.00	40.00	16.20	1.00	17
18	91.00	60.00	37.20	1.00	18
19	168.00	50.00	20.00	1.00	19
20	219.90	135.00	40.90	1.00	20

(continued p. 44)

Mean Time Between Defective Removals(continued):

	OF INPUTS FO DENCE LEVEL =	OR OPTION 3 Y= 0.5	•a*b†X1*c↑%2	
CORRELATION MATE 1.00000 6 0.13621 1 0.08772 -0 0.46002 -0	(IX (Y is la: 0.13621 0.6 0.00000 -0.3 0.27465 1.6 0.53816 -0.3	st column) 38772		
CONFIDENCE OF VE 7.338239574	ARIABLE 2	- NEXT ITERATION ENTERED IS	0.986214257	F =
VARIABLE 2 CONSTANT	COEF 0.943047895 1088.225032	STAND ERI 0.021646353	R NAME COMPLEX	
ANOVA TABLE SOURCE	DF	98	MS	OVERALL F
TOTAL REGRESSION RESIDUAL	19 1 13	27.50176129 7.96481983 19.53694146	7.96481983 1.085385637	7.33823
			28.96112634	
CONFIDENCE OF VI 11.71423952	ARIABLE 1	- NEXT ITERATION ENTERED IS	0.996521100	F =
VARIABLE 1 2 CONSTANT	00EF 1.017222801 0.935472996 302.176393	STAND ER 4.98923E-03 0.017299706	R NAME TECHNO COMPLEX	
ANOVA TABLE SOURCE	DF	SS	MS	OVERALL F
	19 2 17	27.50176129 15.93509575 11.56666554	7.967547874 0.680392091	11.7102
THE OVERALL CON THE PER CENT CF	FIDENCE IS EFROR EXPLAI	0.996517207 HED IS	57.94209171	

Continued p.45)

Mean Time Between Defective Removals(Continued):

		* * · · ·		
CONFIDENCE OF V6 46.32170487	APIABLE 3	NEXT ITERATION ENTERED IS	0.999967668	F =
VARIABLE 1 2 3 CONSTANT	COEF 1.019616009 0.918176142 0.222921617 2144.098142	STAND ER 2.62856E-03 9.44226E-03 0.220530994	R NAME TECHNO COMPLEX UTILITY	
ANOVA TABLE SOURCE	DF	98	MS	OVERALL F
TOTAL REGRESSION RESIDUAL	19 3 16	27.50176129 24.53222364 2.969537644	8.177407881 0.185596103	44.0602
THE OVERALL CON THE PER CENT OF	FIDENCE IS ERROR EXPLAIN	0.999961728 ED IS	89.20237285	5
NO OTHER VARIABI THE AVERAGE % EI THE AVERAGE ERR	RROR IS 3	5.48		
	PARTIAL F STA	T CONF.	NAME	NAME 61AA0 63AA0 63AA0 63AA0 65AA0 65BA0 65BA0 71ZA0 7
1 2 3	54.6180118 9 1.73685646 46.32170437	0.999981256 0.999994668 0.999967668	TECHNO COMPLEX UTILITY	

Bench Check Serviceable Elapsed Time 6428 BSET = .2267(complexity)

	OBSERVATIONS	CROSS	REF:	DATA	FΤ
1	1		636	HOR	
2	2		636	10AF	
3	3		656	RAOR	
4	4		656	TORF	
5	5		746	RBOR	
6	6		741	ACØT	

THE INPUT VARIABLES ARE

	BCS	COMPLE	
N	Y	× 1	H
1	1.50	8.80	1
2	0.60	3.80	2
3	0.40	3.40	3
4	0.40	3.40	4
5	2.30	40.90	5
6	3.30	75,30	6

(continued p.47)

Bench Check Serviceable Elapsed Time(Continued):

---- REGRESSION OF INPUTS FOR OPTION 2 Y=a*X1fb*X2fc -----CONFIDENCE LEVEL = 0.5 CORRELATION MATRIX (Y is lost column) 1.00000 0.95575 0.95575 1.30000 ----- NEXT ITERATION -----CONFIDENCE OF VARIABLE 1 42.22277712 ENTERED IS 0.995816937 F = STAND ERR VARIABLE COEF NAME 0 642784727 0.098921832 COMPLEX 0 226743568 CONSTANT ANOVA TABLE SOURCE $\mathbf{D}^{(i)}$ SS MS OVERALL F TOTAL 5 4.222393756 REGRESSION 3.856998683 3.856998683 42.2227 RESIDUAL 0.365395073 0.091348768 THE OVERALL CONFIDENCE IS 0.995816937 THE PER CENT OF ERROR EXPLAINED IS 91.34625774 NO OTHER VARIABLES CAN BE ADDED THE AVERAGE % ERROR IS 19.38
THE AVERAGE EPROR IS 0.3 0.225599304 OBSERV EST Y TRUE Y ERROR % ERROR NAME 0.58 0.92 1.50 38.83 63AAQR 1 23 0.53 0.07 0.60 10.86 63AAQT 0.50 0.40 -0.10 -24.48 65AAQR 0.40 0.50 65AA0T -0.10 -24.48 2.462.30 -0.16-7.10748B0P 3.65 -0.35 3.30 -10.51 74AC01 PARTIAL SIGNIFICANCE OF VARIABLES IN REGRESSION VAR PARTIAL F STAT CONF. NAME

0.995816937

COMPLEX

43.22277712

1

Bench Check and Repair Elapsed Time

BCRT*= (FIAT + FIXT)/(Efficiency)

where FIAT = Fault Isolate and Test Time =

.3237(complexity)

1 2 3 4 5	OBSERVATIONS 1 2 3 4 5	CROSS	636 636 656 656 746	DATA IAOR IAOT IAOR IAOT IBOR	PT
6	6		746	4COT	

THE INPUT VARIABLES ARE

N	BCFIT Y	COMPLE X 1	X N
1	2.00	8.80	1
2	0.90	3.80	2
3	0.60	3.40	3
4	0.60	3.40	4
5	3.90	40.90	5
É	5.00	75.30	6

see p.29 for description of total equation.ANOVA table for BCRT on p.45.

Bench Check and Repair Elapsed Time(Continued):

---- REGRESSION OF INPUTS FOR OPTION 2 CONFIDENCE LEVEL = 0.5 Y=a*X11b*X21c -----CORPELATION MATRIX (Y is last column) 0.47046 1.00000 0.97046 1.00000 ----- NEXT ITERATION ------CONFIDENCE OF VARIABLE 1 ENTERED IS 0.997590763 F = 64.72737243 VARIABLE COEF STAND ERR NAME 0 665024030 0.082659610 COMPLEX CONSTANT 0 323729658 ANOVA TABLE $\mathbf{D} \in$ SOURCE 88 MS OVERALL F TOTAL 4.383639509 4.12850742 0.255132088 REGRESSION 4.12850743 64,727 1 RESIDUAL 4 0.063783022 THE OVERALL CONFIDENCE IS: 0.997590763 THE PER CENT OF EMROR EMPLAINED IS 94.17990262 NO OTHER VARIABLES CAN BE ADDED THE AVERAGE % ERROR IS 17.34
THE AVERAGE ERROR IS 0.3 0.301959725 TRUE Y OBSERV EST Y ERROR % ERROR HAME 1.37 0.79 0.73 0.73 1 2 2.000.6331.25 **63880P** 0.90**63880T** 12.60 0.11-21.75 -21.75 0.60-0.13 65AA0R 0.60**65AA0T** -0.13 74AB0F 3.900.082.065.00 748C0T ~0.73 -14.64PARTIAL SIGNIFICANCE OF VARIABLES IN REGRESSION PARTIAL F STAT CONF. NAME

0.997590763

COMPLEX

64.72737243

1

1 2 4 5 6 7 8	OBSERVATIONS 1 2 3 4 5 6	CROSS	638 638 658 658 748 748	AOT AOR AOT BO ICO	PT
6	6 7		620	AOP	
ဗ	8 9			AOT AOR	
10	10			IABT	
ii	. 11		718	ene	

THE INPUT VARIABLES ARE

н	ITA Cost	COMPLEX X 1	Н
	16727.00	8.80	1
2	7169.00	3.80	2
3	4699.00	3.40	3
4	4699.00	3.40	4
5	19987.00	40,80	5
ő	21264.00	75.30	6
7	16253.00	7.10	₹
8	6965.00	3.00	8
9	11684.00	12.20	ą.
10	7790.00	8.10	10
11	7668 .0 0	4.50	11

(continued p. 51)

Integrated Test Adaptor Cost(Continued): ---- REGRESSION OF INPUTS FOR OPTION 2 Y=a*X1fb*X2fc -----CONFIDENCE LEVEL = 0.5 CORRELATION MATRIX (Y is last column) 0.83781 1.000000.837811.00000 ----- NEXT ITERATION ------CONFIDENCE OF VARIABLE 1 ENTERED IS 0.998401447 F = 21.19386615 VARIABLE COEF STAND ERR NAME 0.440124445 0.095602753 COMPLEX CONSTANT 3919.226446 ANOVA TABLE SOURCE SS MS OVERALL F TOTAL 10 3.087453208 REGRESSION 1 2.16716434 2.16716434 21.1938 0.102254319 RESIDUAL 9 0.920288867 THE OVERALL CONFIDENCE IS 0.998401447 THE PER CENT OF ERPOR EXPLAINED IS 70.19262139 NO OTHER VARIABLES CAN BE ADDED THE AVERAGE % ERROR IS THE AVERAGE ERROR IS 20.90 2320.138875 OBSERV EST Y TRUE Y ERROR % ERROR NAME 10206.82 16727.00 6520.18 1 38.98 63AAOR 115.94 -2017.10 7953.06 7169.00 1.62 **63880T** 4699.00 3 6716.10 -42.93 65AAOR 4699.00 -2017.10-42.93 65AAGT 6716.10 20048.93 26255.71 19987.00 -61.93 -0.31 74AB0 21264.00 -4991.71 -23.4774AC0 9286.67 16253.00 696**6.**33 42.86 620A0R 6356.13 6965.00 62CA0T 608.87 8.74 1178**5.**12 9841.17 11684.00 7790.00 -101.12 -2051.17 9 -0.87 71880R 10 -26.33 71AA0T 7597.93 11 7668.00 70.07 718A0 0.91 PARTIAL SIGNIFICANCE OF VARIABLES IN REGRESSION

CONF.

0.998401447 COMPLEX

NAME

PARTIAL F STAT

21.19386615

VAR

1

Test Software Cost

SWC = 9867 + 1431(complexity)

	OBSERVATIONS	CROSS	REF:	DATA	PT	1
1	1			AAOR		
2	2		631	AAOT		
2	3		651	AAØR		
4	4		65	AAOT		
5	5		74	AB0		
É	6			ACØ		
7	7			CAØR		
8	8			CAOT		
9	9		71	aa0r		
10	1/3		71	AAOT		

THE INPUT VARIABLES ARE

	SM Cost	COMPLE	
И	Y	X 1	N
1	32929.00	8.80	1
2	14113.00	3.80	2
3	15341.00	3.40	3
4	15341.00	3.40	4
5	773 5 8. 0 0	40.80	5.
6	113662.00	75.30	ઈ
7	27150.00	7.10	7
8	11636.00	3.00	8
9	17165.00	12.20	9
10	11444.00	8.10	10

(continued p. 53)

Test Software Cost(Continued): ---- REGRESSION OF INPUTS FOR OPTION 1 Y=a+b*X1+c*X2 ------CONFIDENCE LEVEL = 0.5 COPRELATION MATRIX (Y is last column) 0.97765 1.00000 0.97765 1.00000 CONFIDENCE OF VARIABLE 1 ENTERED IS 0.99997493 F = 173.0123672 VARIABLE COEF STAND ERR NAME 1431.396598 9867.030441 108.8231784 COMPLEX CONSTANT ANOVA TABLE D. 88 MS OVERALL F SOURCE TOTAL 9 1.06567E+10 REGRESSION 1.01857E+10 1.01857E+10 1 173.01 RESIDUAL 8 470981184.5 58872648.07 THE OVERALL CONFIDENCE IS 0.99997493 THE PER CENT OF ERROR EXPLAINED IS 95.58041248 NO OTHER VARIABLES CAN BE ADDED THE AVERAGE % ERROR IS 25.81 THE AVERAGE ERROR IS 55 5578.032766 TRUE Y % ERROR NAME OBSERV EST Y ERROR 22463.32 15306.34 10465.68 32929.00 31.78 63AABR 14113.00 15341.00 -1193.34 -8.46 3.96 63AAØT 14733.78 65AA0R 607.22 14733.78 15341.00 607.22 3.96 65AA0T 11.75 -3.51 77358.00 9089.99 5 68268.01 74AB0 -3989.19 117651.19 113662.00 74AC0 20029.95 27150.00 7120.05 26.22 620A0R 14161.22 27330.07 -2525.22 -10165.07 **62CA0T** 8 11636.00 -21.70 17165.00 -59.22 71AAOR 21461.34 11444.00 -10017.34 71880T 10 PARTIAL SIGNIFICANCE OF VARIABLES IN REGRESSION VAR PHRTIAL F STAT NAME CONF. 173,0123675 0.99997493 COMPLEX

Technical Order Pages
4366
TOP = 69.64(complexity)

THE I	NPUT	VARI	ABL	ES.	ARE.
-------	------	------	-----	-----	------

N	PAGES Y	COMPLI X 1	EX N
1	88.00	3.10	1
2	140.00	3.40	2
3	102.00	3.60	3
4	140.00	3.40	4
5	206.00	7.20	5
ϵ	164.00	9.00	6
7	328.00	34.00	7
8	362.00	78.00	8
9	371.00	41.00	9
10	550.00	75.00	10

(continued p. 55)

Technical Order Pages (Continued):

```
---- PEGRESSION OF INPUTS FOR OPTION 2 Y=a*X1fb*X2fc -----
          CONFIDENCE LEVEL = 0.5
COPRELATION MATRIX (Y is lost column)
1.00000 0.95300
    0.95300
               1.00000
----- NEXT ITERATION ------
CONFIDENCE OF VARIABLE 1 ENTERED IS 0.999901066 F =
 79.15527883
                                     STAND ERR
                                                   NAME
VARIABLE
                    COEF
                0.436556126
                               0.049068202 COMPLEX
                69.64481206
CONSTANT
ANOVA TABLE
                                  SS
                DF
                                                     MS OVERALL
SOURCE
                9
                                3.412347387
TOTAL
REGRESSION
                                3.099127357
                                               3.099127357
                                                              79.15
                1
                                               0.039152504
PESIDUAL
                                0.313220030
THE OVERALL CONFIDENCE IS
                               0.999901066
THE PER CENT OF ERROR EXPLAINED IS:
                                              90.82098055
NO OTHER VARIABLES CAN BE ADDED
THE AVERAGE % ERROR IS 16.17
THE AVERAGE ERROR IS 36
                                36.50707500
                                               % ERROR
                       TRUE Y
                                   ERROR
                                                           HAME
OBSERV
            EST Y
         114.13
                                  -26.13
                     38.00
                                              -29.69
  1
                                              15.12
         118.83
                    140.00
                                  21.17
                    102.00
                                  -19.83
                                              -19.44
  3
         121.83
                     140.00
                                  21.17
                                               15.12
         118.83
         164.88
                      206.00
                                  41.12
                                               19.96
         181.75
324.69
                      164.00
                                  -17.75
                                               -10.82
                      328.00
                                    3.31
                                                1.01
                      362.00
          466.55
                                 -104.55
                                               -28.88
  8
          352.34
                                                5.03
  9
                      371.00
                                   13.66
         458.63
                      550.00
PARTIAL SIGNIFICANCE OF VARIABLES IN REGRESSION VAR PARTIAL F STAT CONF.
                                                   NAME
                 79.15527886
                                0.999901066 COMPLEX
 1
```